

Lecture 4

Submanifold reconstruction

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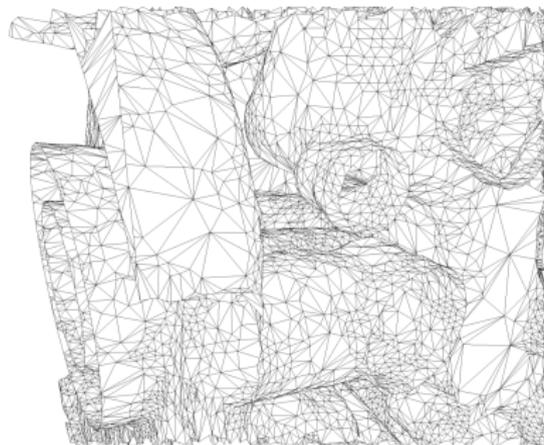
Winter School on Computational Geometry and Topology
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Outline

- 1 The reconstruction problem
- 2 Distance functions and homotopy reconstruction
- 3 Delaunay-type simplicial complexes and homeomorphic submanifold reconstruction
- 4 Mesh generation of surfaces

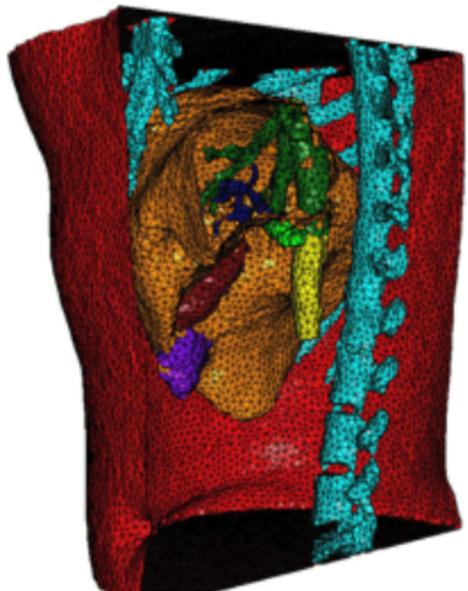
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Reconstructing surfaces from point clouds

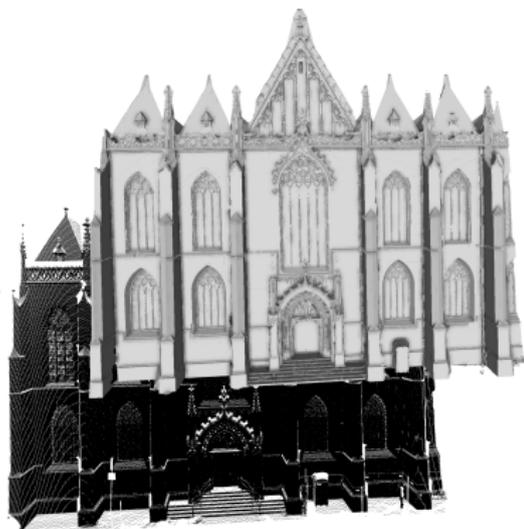
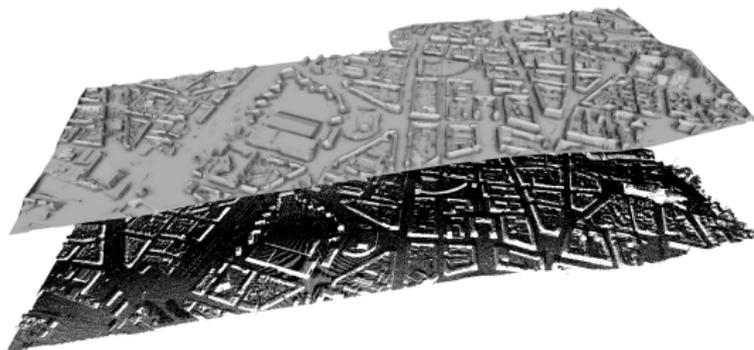


One can reconstruct a surface from 10^6 points within 1mn

[CGAL]

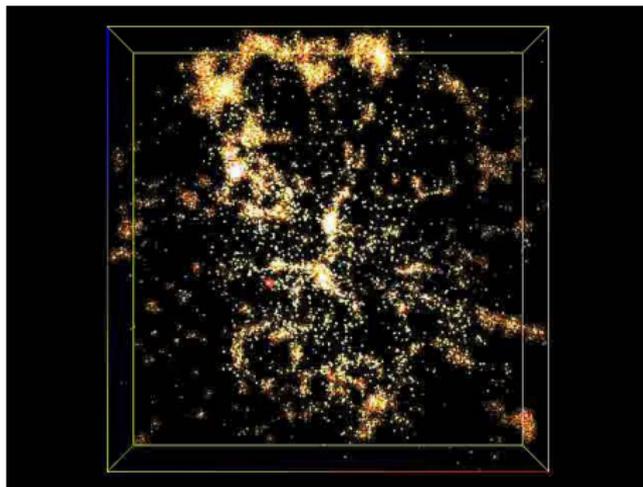


CGAL-mesh, Titane
GeometryFactory, Acute3D



Geometric data analysis

Images, text, speech, neural signals, GPS traces,...

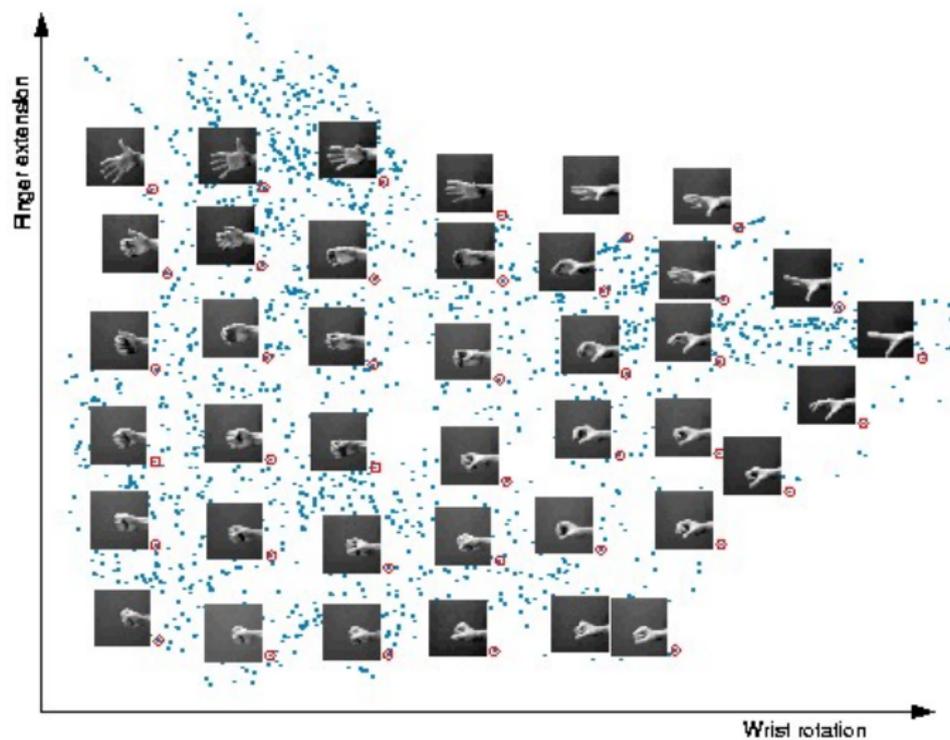


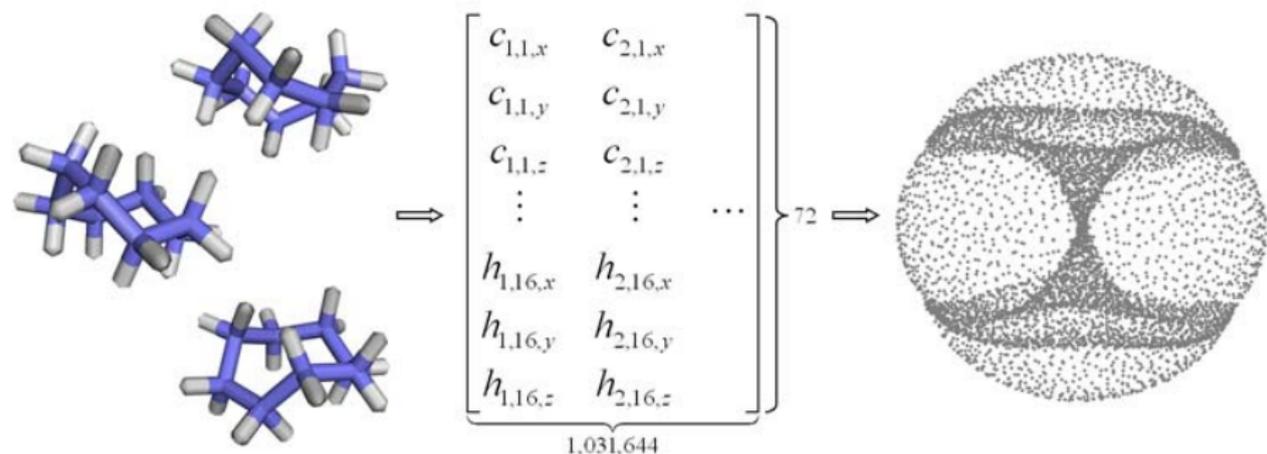
Geometrisation : Data = points + distances between points

Hypothesis : Data lie close to a structure of
“small” intrinsic dimension

Problem : Infer the structure from the data

Dimensionality reduction





- Each conformation is represented as a point in \mathbb{R}^{72} (\mathbb{R}^{24} when neglecting the H atoms)
- The intrinsic dimension of the conformation space is 2
- The geometry of C_8H_{16} is highly nonlinear

Image manifolds

An image with 10 million pixels

→ a point in a **space of 10 million dimensions!**



camera : 3 dof
light : 2 dof

The image-points lie close to a structure of **intrinsic dimension 5** embedded in this **huge ambient space**

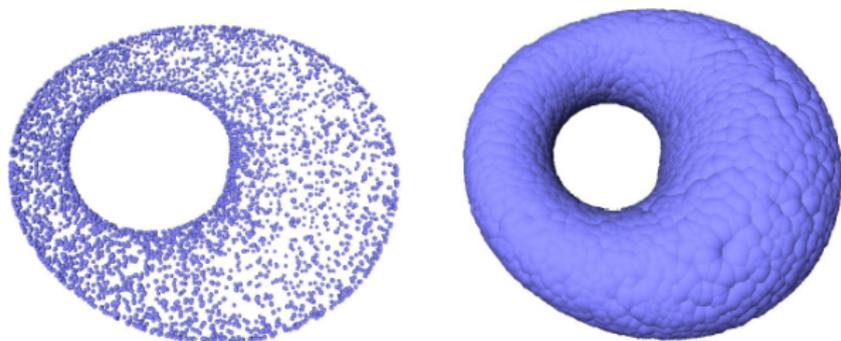
Motion capture



Typically $N = 100$, $D = 100^3$, $d \leq 15$

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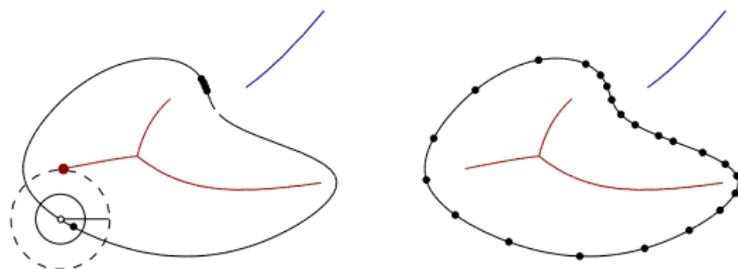
Distance to a compact K : $d_K : x \rightarrow \inf_{p \in K} \|x - p\|$



Stability

If the data points C are ε -close (Hausdorff) to the geometric structure K , the topology and the geometry of the offsets $K_r = d_K^{-1}([0, r])$ and $C_r = d_C^{-1}([0, r])$ are close for $r \in [\Omega(\varepsilon), \text{Reach}(K) - \Omega(\varepsilon)]$

The **medial axis** of K is the set of points of the complement of K with at least two closest points on K



A finite point set $P \subseteq K$ is an ε -**net** of K if

- 1 **Covering:** $\forall x \in K, d(x, P) \leq \varepsilon \text{lfs}(x)$
- 2 **Packing:** $\forall p, q \in P, \|p - q\| \geq \eta_0 \varepsilon \max(\text{lfs}(p), \text{lfs}(q))$ for some cst η_0

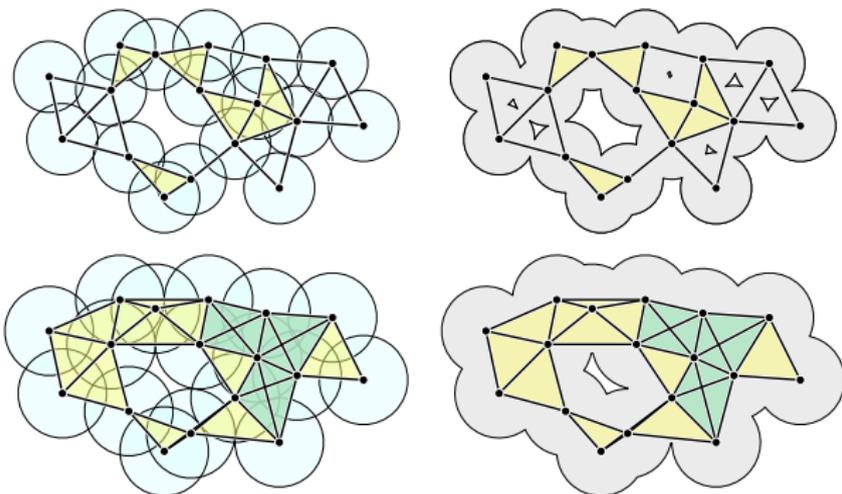
lfs denotes the distance from x to the medial axis of \mathbb{M}

Properties of lfs

- lfs is 1-Lipschitz : $|\text{lfs}(x) - \text{lfs}(y)| \leq \|x - y\|$
- $\text{lfs}(x)$ is small where the curvature is large and where the thickness of \mathbb{M} is small
- $\text{lfs} > 0$ if S is $C^{1,1}$
i.e. normals exist everywhere and the normal field is Lipschitz

$\inf_{x \in \mathbb{M}} \text{lfs}(x)$ is called the **reach** of \mathbb{M}

Distance functions and simplicial complexes



Nerve theorem (Leray)

The nerve of the balls (Cech complex) and the union of balls have the same homotopy type

Some remarks and questions

From continuous to discrete and back:



- + The topology of a compact set K can be computed from the Čech complex of a sample P of K
- The Čech complex is **huge** ($O(n^d)$) and very difficult to compute
- The Čech complex is in general **not** homeomorphic to K (a **triangulation** of K)
- The Čech complex **cannot be realized** in general in the same space as K
- ~ Replace the α -Čech complex by the α -complex (less big and embedded)

Looking for small and faithful simplicial complexes

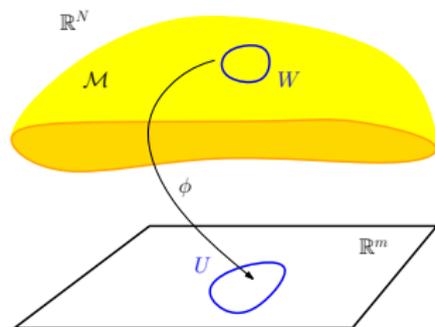
Need to compromise

- Size of the complex
 - ▶ can we capture the intrinsic dimensionality ?
- Efficiency of the construction algorithms and of the representations
 - ▶ can we avoid the exponential dependence on d ?
 - ▶ can we minimize the number of simplices ?
- Quality of the approximation
 - ▶ Homotopy type & homology (RIPS complex, persistence)
 - ▶ Homeomorphism (Delaunay-type complexes)

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Submanifolds of \mathbb{R}^d

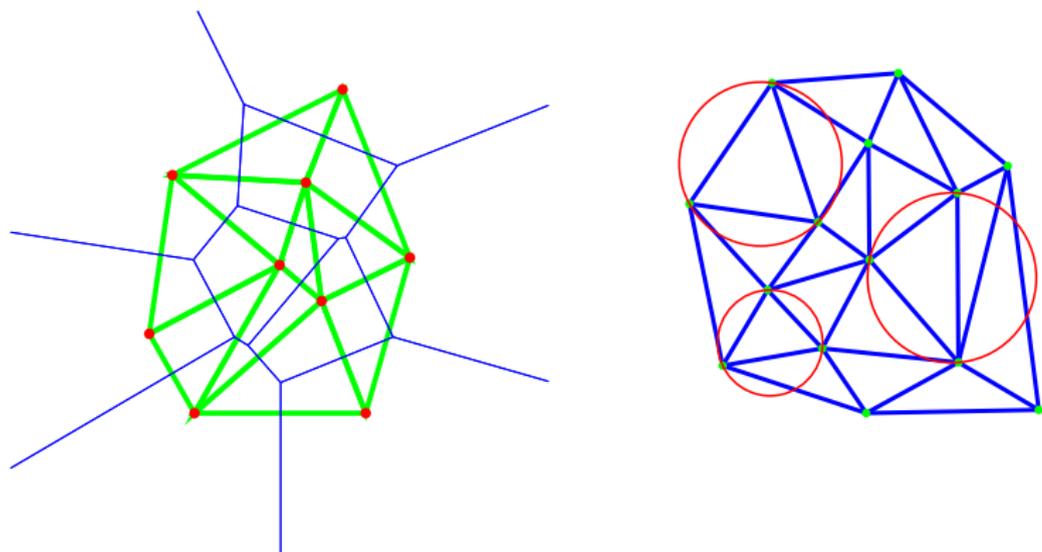
A compact subset $M \subset \mathbb{R}^d$ is a submanifold without boundary of **intrinsic** dimension $k < d$, if any $p \in M$ has an open (topological) k -ball as a neighborhood in M



A **curve** is a 1-dimensional submanifold

A **surface** is a 2-dimensional submanifold

Voronoi diagram and Delaunay complex



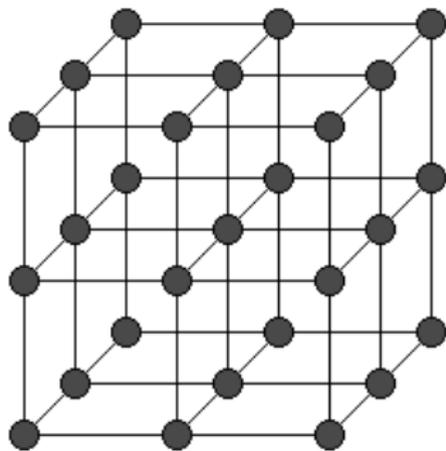
Delaunay complex : $\text{Del}(P) = \text{nerve of Vor}(P)$

Equivalently, $\text{Del}(P)$ is the collection of simplices with an **empty** circumscribing ball

The curses of Delaunay triangulations in higher dimensions

- Restricted to Euclidean space (see otherwise Mael's talk)
- Computing DT is restricted to low dimensions
(The number of simplices grows exponentially with d even if the vertices lie on a curve !)
- 3 and higher dimensional Delaunay triangulations are not thick even if the vertices are well-spaced

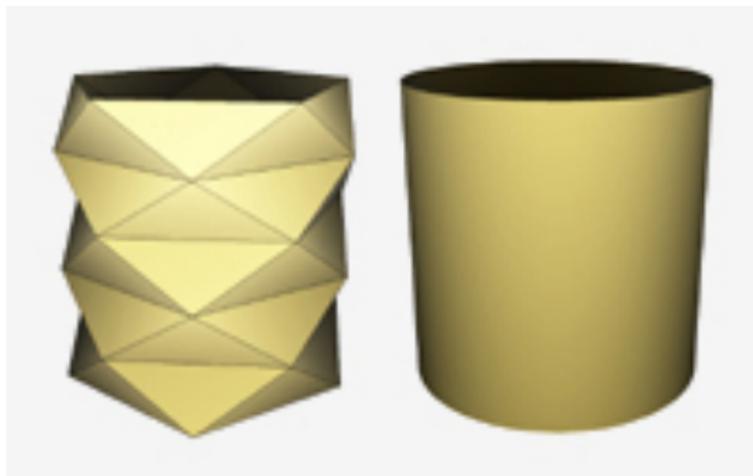
3D Delaunay Triangulations are not thick even if the vertices are well-spaced



- Each square face can be circumscribed by an empty sphere
- This remains true if the grid points are slightly perturbed therefore creating thin simplices

Badly-shaped simplices

Badly-shaped simplices lead to bad geometric approximations

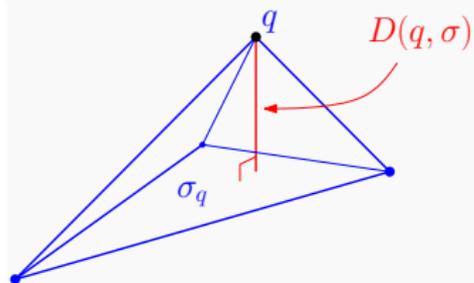


Bad consequences in rendering, numerical simulations, volume calculation and more...

see also [Cairns], [Whitehead], [Munkres], [Whitney]

Simplex quality

Altitudes



If σ_q , the face opposite q in σ is protected, The *altitude* of q in σ is

$$D(q, \sigma) = d(q, \text{aff}(\sigma_q)),$$

where σ_q is the face opposite q .

Definition (Thickness

[Cairns, Whitney, Whitehead et al.])

The *thickness* of a j -simplex σ with diameter $\Delta(\sigma)$ is

$$\Theta(\sigma) = \begin{cases} 1 & \text{if } j = 0 \\ \min_{p \in \sigma} \frac{D(p, \sigma)}{j \Delta(\sigma)} & \text{otherwise.} \end{cases}$$

Tangent space approximation

Lemma

[Whitney 1957]

If σ is a j -simplex whose vertices all lie within a distance h from a hyperplane $H \subset \mathbb{R}^d$, then

$$\sin \angle(\text{aff}(\sigma), H) \leq \frac{2jh}{D(\sigma)} = \frac{2h}{\Theta(\sigma) \Delta(\sigma)}$$

Corollary

If σ is a j -simplex, $j \leq k$, $\text{vert}(\sigma) \subset \mathbb{M}$, $\Delta(\sigma) \leq 2\varepsilon \text{rch}(\mathbb{M})$

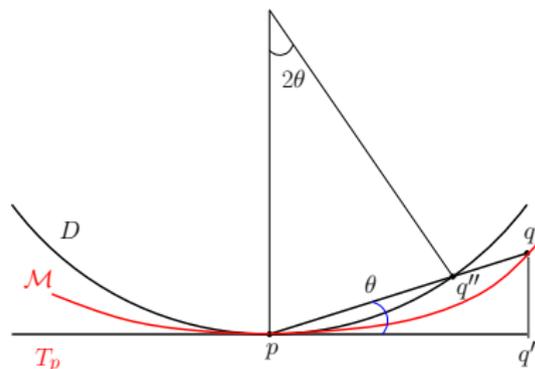
$$\forall p \in \sigma, \quad \sin \angle(\text{aff}(\sigma), T_p) \leq \frac{2\varepsilon}{\Theta(\sigma)}$$

($h \leq \frac{\Delta(\sigma)^2}{2 \text{rch}(\mathbb{M})}$ by the Chord Lemma)

Chord Lemma

Let x and y be two points of \mathbb{M} . We have

- 1 $\sin \angle(xy, T_x) \leq \frac{\|x-y\|}{2 \operatorname{rch}(\mathbb{M})}$;
- 2 the distance from y to T_x is at most $\frac{\|x-y\|^2}{2 \operatorname{rch}(\mathbb{M})}$.



$$\|x - y\| \geq \|x - y''\| \\ = 2 \operatorname{rch}(\mathbb{M}) \sin \angle(xy, T_x)$$

$$\|y - y'\| = \|x - y\| \sin \angle(xy, T_x) \\ \leq \frac{\|x - y\|^2}{2 \operatorname{rch}(\mathbb{M})}$$

The curses of Delaunay triangulations in higher dimensions

- Restricted to Euclidean space (see otherwise Mael's talk)

⇒ Define local Euclidean triangulations

- Computing DT is restricted to low dimensions

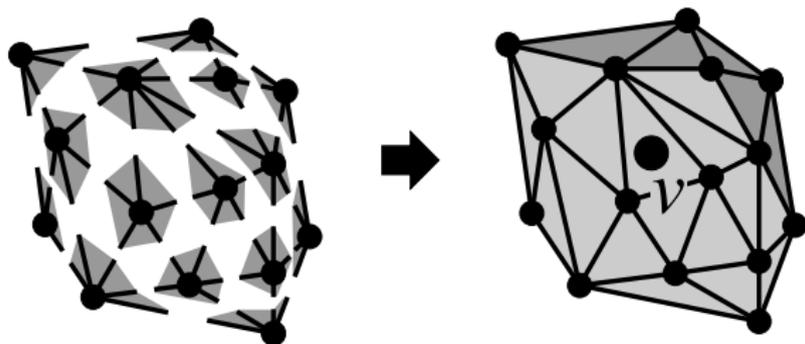
(The number of simplices grows exponentially with d even if the vertices lie on a curve !)

⇒ Exploit the fact that \mathbb{M} has an intrinsic dimension $k \ll d$?

- 3 and higher dimensional Delaunay triangulations are not thick even if the vertices are well-spaced

⇒ Remove flat simplices

Towards Delaunay triangulation of manifolds

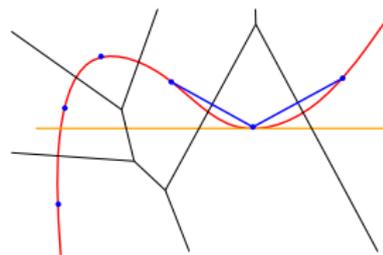
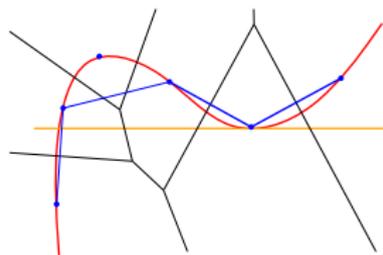
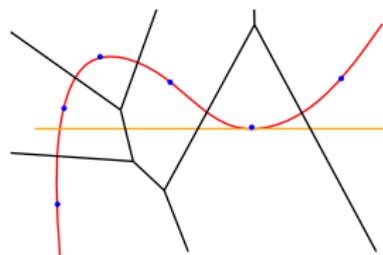


- 1 Construct **local** Delaunay triangulations
- 2 Ensure that the triangulations are **stable** under small perturbation
 \Rightarrow a simplex belongs to the stars of all its vertices
- 3 Glue all local triangulations into a single **triangulated manifold**

The assumptions

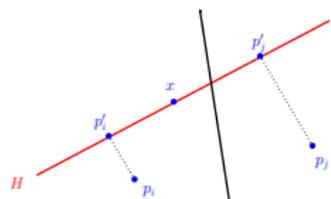
- \mathbb{M} is a differentiable submanifold of **positive reach** of \mathbb{R}^d
- The dimension k of \mathbb{M} is **small**
- \mathcal{P} is an **ε -net** of \mathbb{M} for a small enough ε
- We assume that we know the tangent space T_p at each $p \in \mathcal{P}$

Local triangulation : $\text{Del}_{T_p}(\mathcal{P})$



Constructing $\text{Del}_{T_p}(\mathcal{P})$

Given a d -flat $H \subset \mathbb{R}^d$, $\text{Vor}(\mathcal{P}) \cap H$ is a **weighted** Voronoi diagram in H



$$\|x - p_i\|^2 \leq \|x - p_j\|^2$$

$$\Leftrightarrow \|x - p'_i\|^2 - \|p_i - p'_i\|^2 \leq \|x - p'_j\|^2 - \|p_j - p'_j\|^2$$

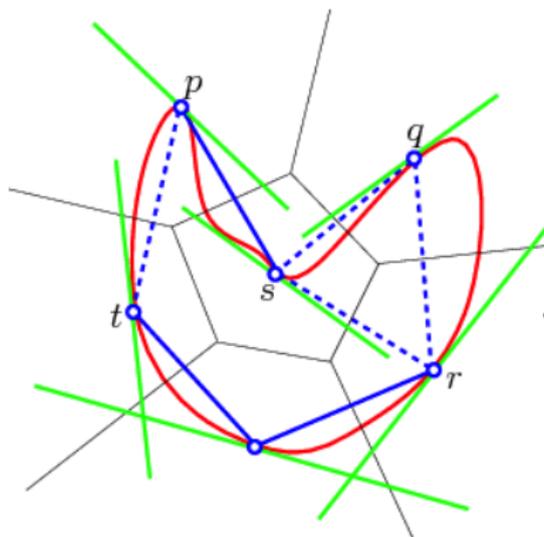
Corollary: construction of Del_{T_p}

$$\psi_p(p_i) = (p'_i, -\|p_i - p'_i\|^2) \quad (\text{weighted point})$$

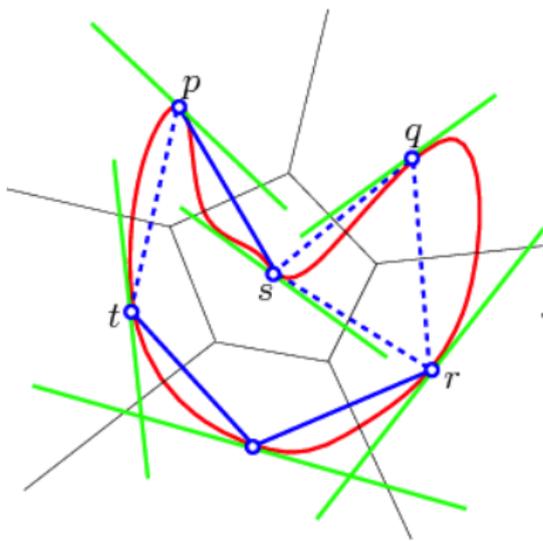
- 1 project \mathcal{P} onto T_p which requires $O(Dn)$ time
- 2 construct $\text{star}(\psi_p(p_i))$ in $\text{Del}(\psi_p(p_i)) \subset T_p$
- 3 $\text{star}(p_i) \approx \text{star}(\psi_p(p_i))$ (isomorphic)

The tangential Delaunay complex

[Freedman 2002], [B.& Flottoto 2004], [B. Ghosh 2014]

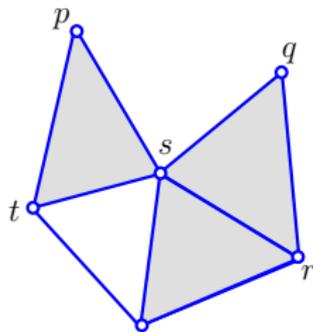
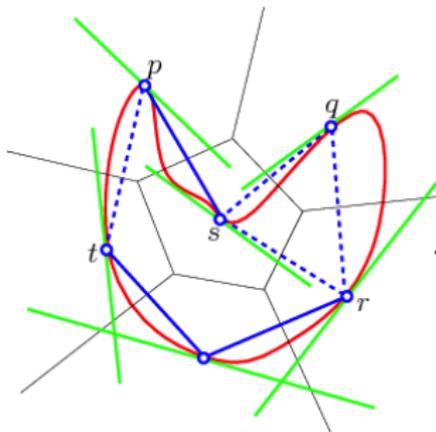


- 1 Construct the star of $p \in \mathcal{P}$ in the Delaunay triangulation $\text{Del}_{T_p}(\mathcal{P})$ of \mathcal{P} restricted to T_p
- 2 $\text{Del}_{TM}(\mathcal{P}) = \bigcup_{p \in \mathcal{P}} \text{star}(p)$



- + $\text{Del}_{TM}(\mathcal{P}) \subset \text{Del}(\mathcal{P})$
- + $\text{star}(p)$, $\text{Del}_{T_p}(\mathcal{P})$ and therefore $\text{Del}_{TM}(\mathcal{P})$ can be computed without computing $\text{Del}(\mathcal{P})$
- $\text{Del}_{TM}(\mathcal{P})$ is **not** necessarily a triangulated manifold

Inconsistent configurations



Definition

$$\phi = [p_1, p_2, \dots, p_{k+2}]$$

$$\tau = \phi \setminus \{p_l\}$$

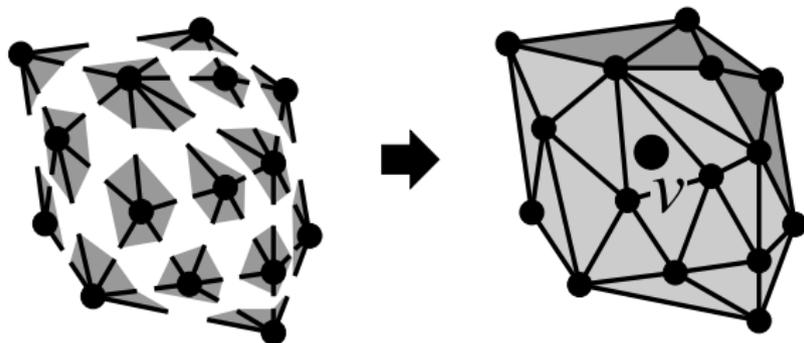
ϕ is an inconsistent configuration witnessed by $p_i, p_j, p_l \in \phi$ if

- $\tau \in \text{starp}_i \iff T_{p_i} \cap \text{Vor}(\tau) \neq \emptyset$
- $\tau \notin \text{starp}_j \iff T_{p_j} \cap \text{Vor}(\tau) = \emptyset$
- $\text{Vor}(p_l)$ is the first Voronoi cell whose interior is hit by $[c_{p_i}(\tau) \rightarrow c_{p_j}(\tau)]$

An IC is a $(k + 1)$ -simplex of $\text{Del}(\mathbb{M})$

Reconstruction of smooth submanifolds

- 1 For each vertex v , compute the star $\text{star}(v)$ of v in $\text{Del}_p(\mathcal{P})$
- 2 Remove inconsistencies among the stars by weighting the points
- 3 Glue the stars to obtain a triangulation of \mathcal{P}



Manifold reconstruction algorithm

Algorithm 1 `Manifold_reconstruction`($P = \{p_0, \dots, p_n\}, \eta_0$)

// Initialization

for $i = 1$ to n **do**

calculate the local neighborhood $LN(p_i)$

for $i = 1$ to n **do**

$\omega(p_i) \leftarrow 0$

Build the full unweighted complex $K^\omega(P) = \text{Del}_{TM}^\omega(P) \cup IC$

// Weight assignment to remove inconsistencies

for $i = 1$ to n **do**

$\omega(p_i) \leftarrow \mathbf{weight}(p_i, \omega)$

 update $K^\omega(P)$ // (locally in $LN(p_i)$)

output : $\hat{M} \leftarrow \text{Del}_{TM}^\omega(P)$

Hypotheses

- \mathbb{M} is a differentiable submanifold of **positive reach** of dim. $k \subset \mathbb{R}^d$
- \mathcal{P} is an **ε -net** of \mathbb{M} for a small enough ε

Theorem

Under the Hypotheses, the algorithm terminates and $\hat{\mathbb{M}}$ contains no inconsistent configurations

\hat{M} is a PL simplicial k -manifold

Lemma

Let P be an ε -sample of a manifold M and let $p \in P$. The link of any vertex p in \hat{M} is a topological $(k - 1)$ -sphere

Proof :

1. Since \hat{M} contains no inconsistencies, the star of any vertex p in \hat{M} is identical to $\text{star}p$, the star of p in $\text{Del}_p(P)$
2. $\text{Del}_p(P) \subset \mathbb{R}^d \approx \text{Del}(\psi_p(P)) \subset T_p \Rightarrow \text{star}p \approx \text{star}_p(p)$
3. $\text{star}_p(p)$ is a k -dimensional triangulated topological ball (general position)
4. p cannot belong to the boundary of $\text{star}_p(p)$
(the Voronoi cell of $p = \psi_p(p)$ in $\text{Vor}(\psi_p(P))$ is bounded)

Guarantees

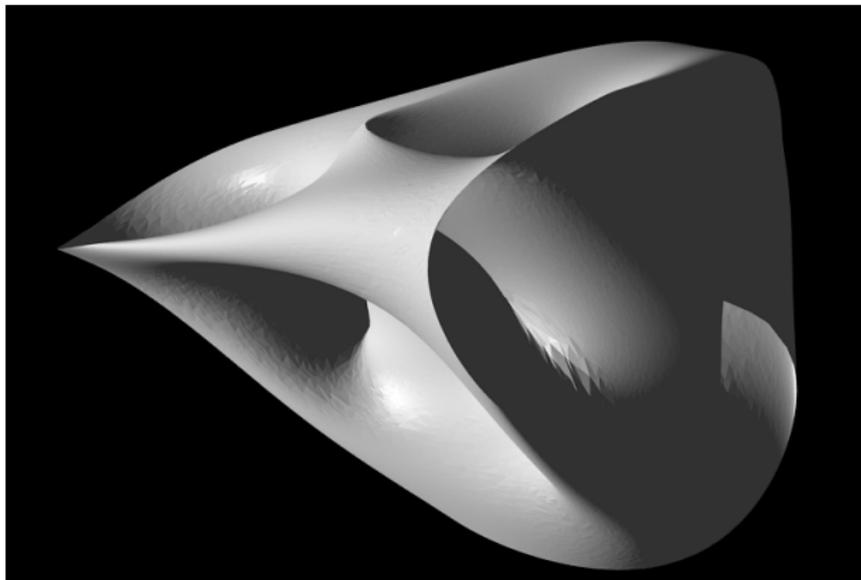
Approximation guarantees

- $\hat{\mathbb{M}}$ is a PL simplicial k -manifold
- $\hat{\mathbb{M}} \subset \text{tub}(\mathbb{M}, O(\varepsilon^2)\text{rch}(\mathbb{M}))$
- The angles between the facets and the tangent spaces of \mathbb{M} are $O(\varepsilon)$
- $\hat{\mathbb{M}}$ is **homeomorphic** to \mathbb{M}

Complexity of the algorithm

- No d -dimensional data structure \Rightarrow **linear in d**
- **exponential in k**

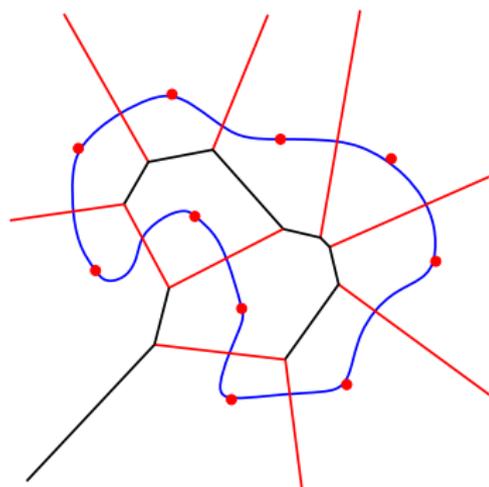
Reconstructing a Riemannian surface in \mathbb{R}^8



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Definition

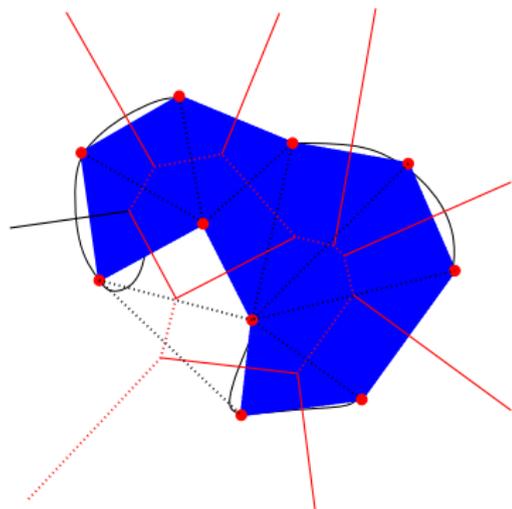
The **restricted Delaunay triangulation** $\text{Del}_X(\mathcal{P})$ to $X \subset \mathbb{R}^d$ is the nerve of $\text{Vor}(\mathcal{P}) \cap X$



If \mathcal{P} is an ε -sample, any ball centered on X that circumscribes a facet f of $\text{Del}_X(\mathcal{P})$ has a radius $\leq \varepsilon \text{ lfs}(c_f)$

Definition

The **restricted Delaunay triangulation** $\text{Del}_X(\mathcal{P})$ to $X \subset \mathbb{R}^d$ is the nerve of $\text{Vor}(\mathcal{P}) \cap X$



Delaunay triang. restricted to surfaces

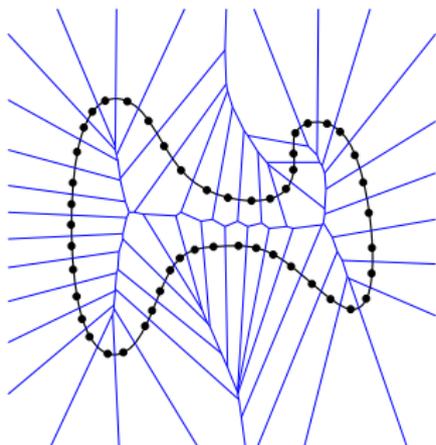
[Amenta et al. 1998-], [B. & Oudot 2005]

If

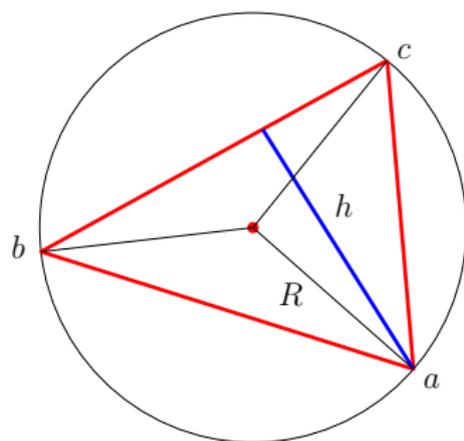
- $\mathcal{S} \subset \mathbb{R}^3$ is a compact **surface** of positive reach without boundary
- \mathcal{P} is an ε -net, ε small enough

then

- $\text{Del}_{|\mathcal{S}}(\mathcal{S})$ provides good estimates of normals
- There exists a **homeomorphism**
 $\phi : \text{Del}_{|\mathcal{S}}(\mathcal{P}) \rightarrow \mathcal{S}$
- $\sup_x (\|\phi(x) - x\|) = \mathcal{O}(\varepsilon^2)$



Thickness and angle bounds are automatic for triangles



$$h = ab \times \sin b = \frac{ab \times ac}{2R}$$

$$\Theta(abc) = \frac{h}{2\Delta} \geq \frac{h}{4R} \geq \frac{\phi_0^2}{32R} \geq \frac{\phi_0^2}{32\epsilon \text{rch}(\mathcal{S})}$$

$$\sin b \geq \frac{\phi_0}{2\epsilon \text{rch}(\mathcal{S})}$$

Surface mesh generation by Delaunay refinement

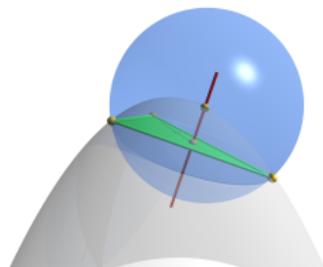
[Chew 1993, B. & Oudot 2003]

$\phi : S \rightarrow \mathbb{R}$ = Lipschitz function

$\forall x \in S, 0 < \phi_0 = \bar{\eta}_0 \varepsilon \text{rch}(S) \leq \phi(x) < \varepsilon \text{lfs}(x)$

ORACLE : For a facet f of $\text{Del}_{|S}(\mathcal{P})$,
return c_f , r_f and $\phi(c_f)$

A facet f is **bad** if $r_f > \phi(c_f)$



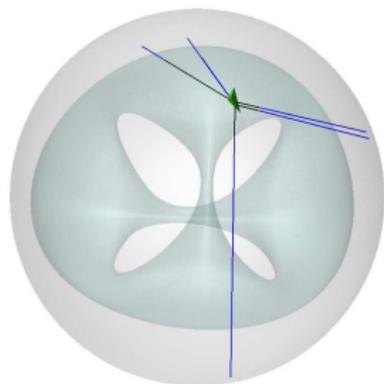
Algorithm

INIT compute an initial (small) sample $\mathcal{P}_0 \subset S$

REPEAT IF f is a bad facet
 insert_in_Del3D(c_f),
 update \mathcal{P} and $\text{Del}_{|S}(\mathcal{P})$

UNTIL no bad facet remains

The meshing algorithm in action

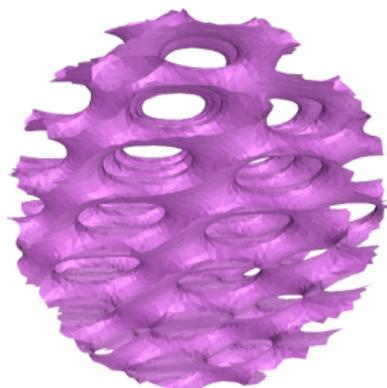
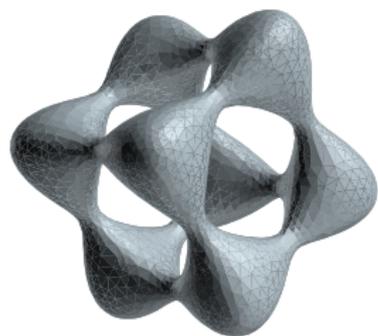


Applications

- Implicit surfaces $f(x, y, z) = 0$
- Isosurfaces in a 3d image (Medical images)
- Triangulated surfaces (Remeshing)
- Point sets (Surface reconstruction)

see cgal.org, CGALmesh project

Results on smooth implicit surfaces



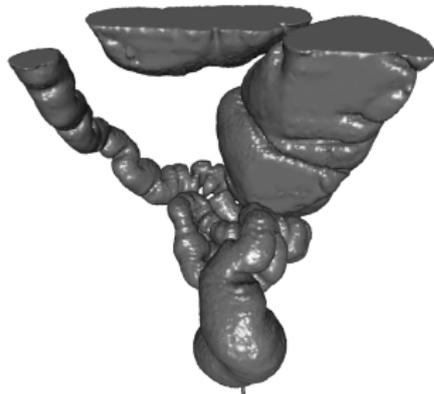
Meshing 3D domains

Input from segmented 3D medical images

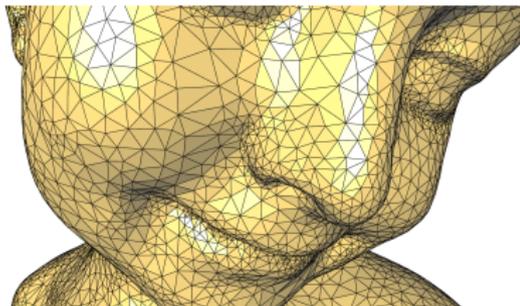
[INSERM]



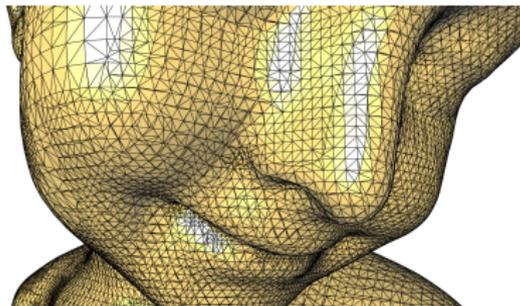
[SIEMENS]



Comparison with the Marching Cube algorithm



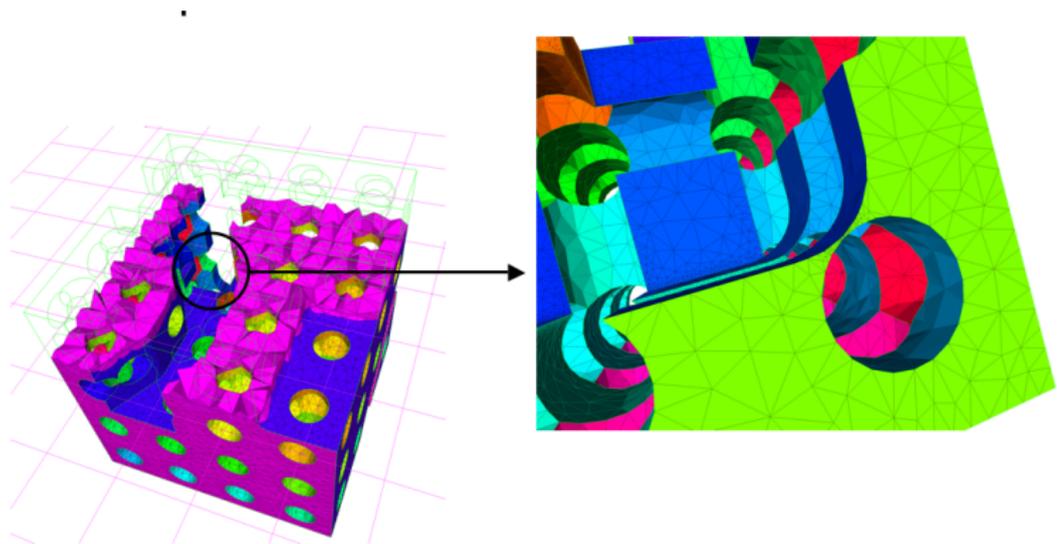
Delaunay refinement



Marching cube

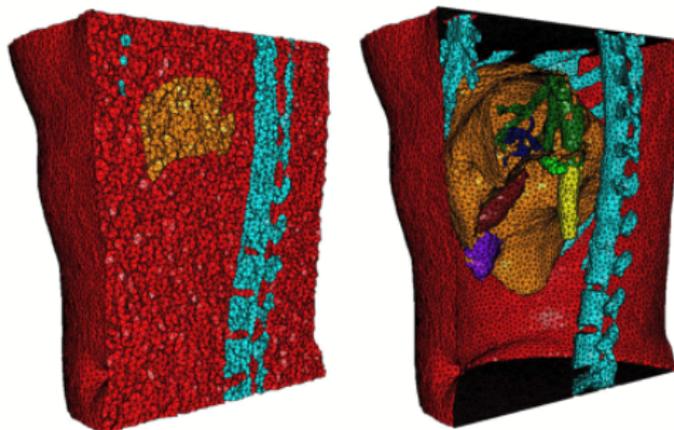
Meshing with sharp features

A polyhedral example



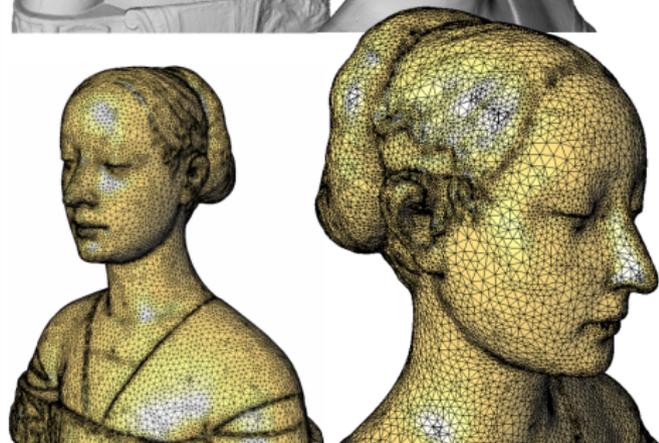
Meshing 3D multi-domains

Input from segmented 3D medical images [IRCAD]



Size bound (mm)	vertices nb	facets nb	tetrahedra nb	CPU Time (s)
16	3,743	3,735	19,886	0.880
8	27,459	19,109	159,120	6.97
4	199,328	76,341	1,209,720	54.1
2	1,533,660	311,420	9,542,295	431

Surface reconstruction from unorganized point sets



Courtesy of P. Alliez